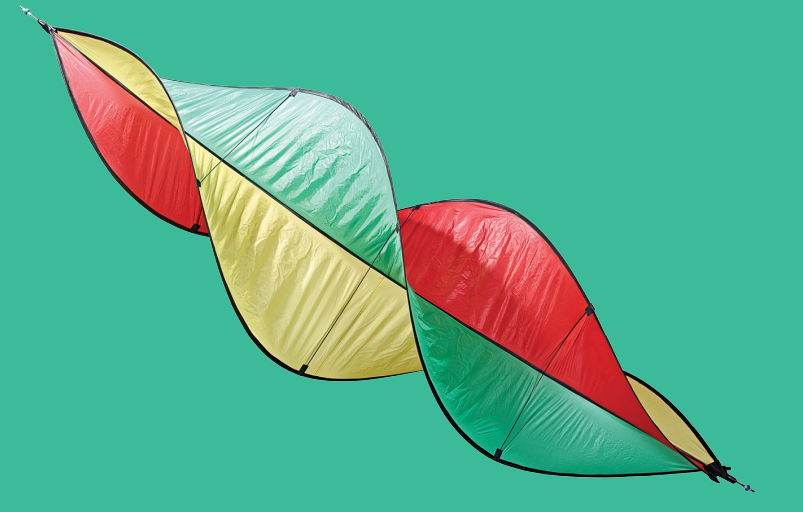


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Embedded & Artificially intelligent ViSion Engineering



Efficient Multiclass Object Detection: Detecting Pedestrians and Bicyclists in a Truck's Blind Spot Camera

Problem & Goal

The Blind spot zone:

- Zone around truck in which driver has no/limited view
- Each year approximately 1300 casualties in Europe

Existing systems:

- 2 types: active and passive



- Passive systems** require truck driver interpretation e.g. mirrors, camera system → not adjusted or used correctly



- Active systems** automatically generate an alarm e.g. ultrasonic or infrared systems → cannot distinguish persons vs static objects (traffic signs)

Goal: Develop a system that automatically detects vulnerable road users (VRUs) in blind spot camera systems

Advantages of such a system:

- Active driver independent
- Automatic detection of vulnerable road users
- Automatic warning on monitor
- Easy to integrate in existing camera systems
- Reliable, provides implicit interpretation



We target **pedestrians** and **bicyclists** (involved mostly in these type of accidents)

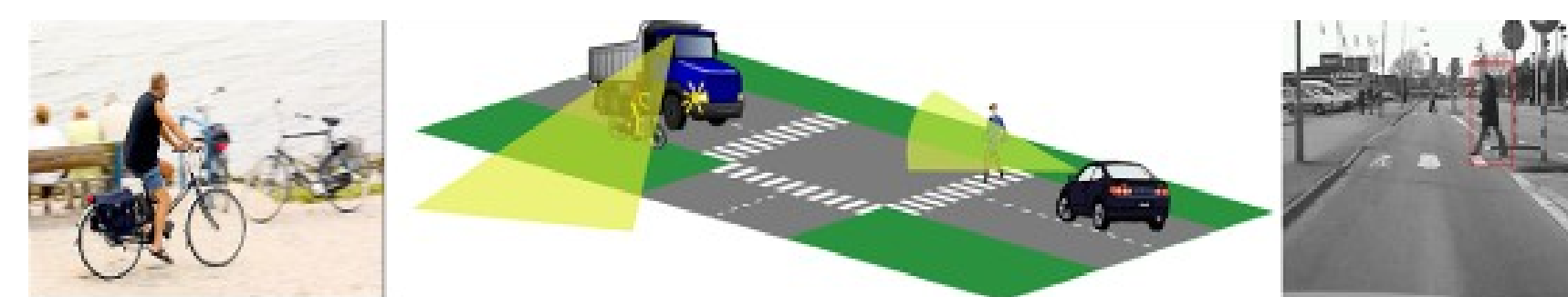


Challenges:

- Hard **real-time**, motion blur, **highly dynamical background**
- Vulnerable road users are a very **diverse class**:



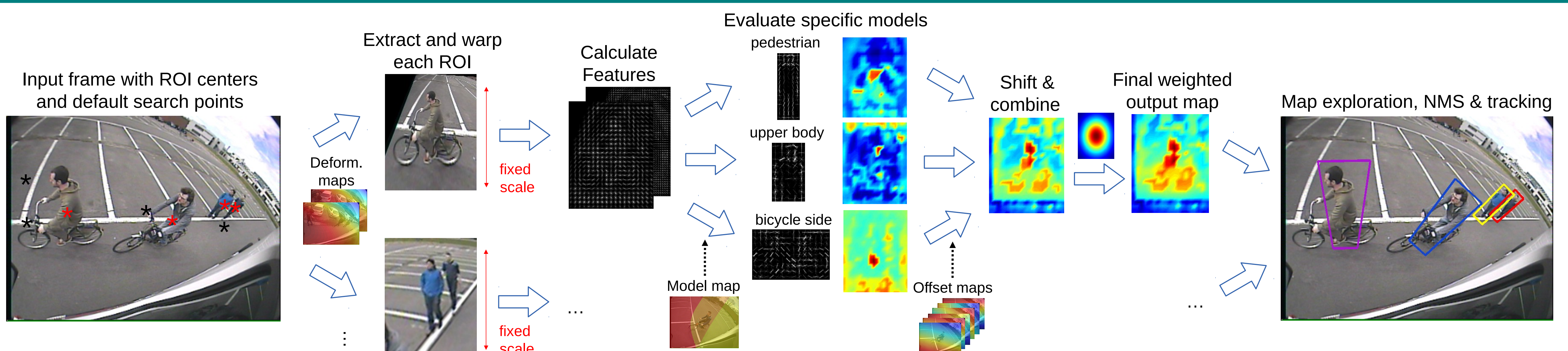
- Blind spot camera: **wide-angle lens & specific viewing direction**



Output detections



Algorithmic Approach



Algorithmic approach overview:

- Traditional object detectors: *search over full scale-space pyramid*: not feasible here
- VRUs appear under various rotations and scales
- Exact **transformation** only depends on **position in image**
- First **warp each ROI** to an undistorted, upright and fixed height image patch
- Run **multiple detection models** with feature sharing
- Finally, **combine probability maps** and integrate in **tracking-by-detection framework**

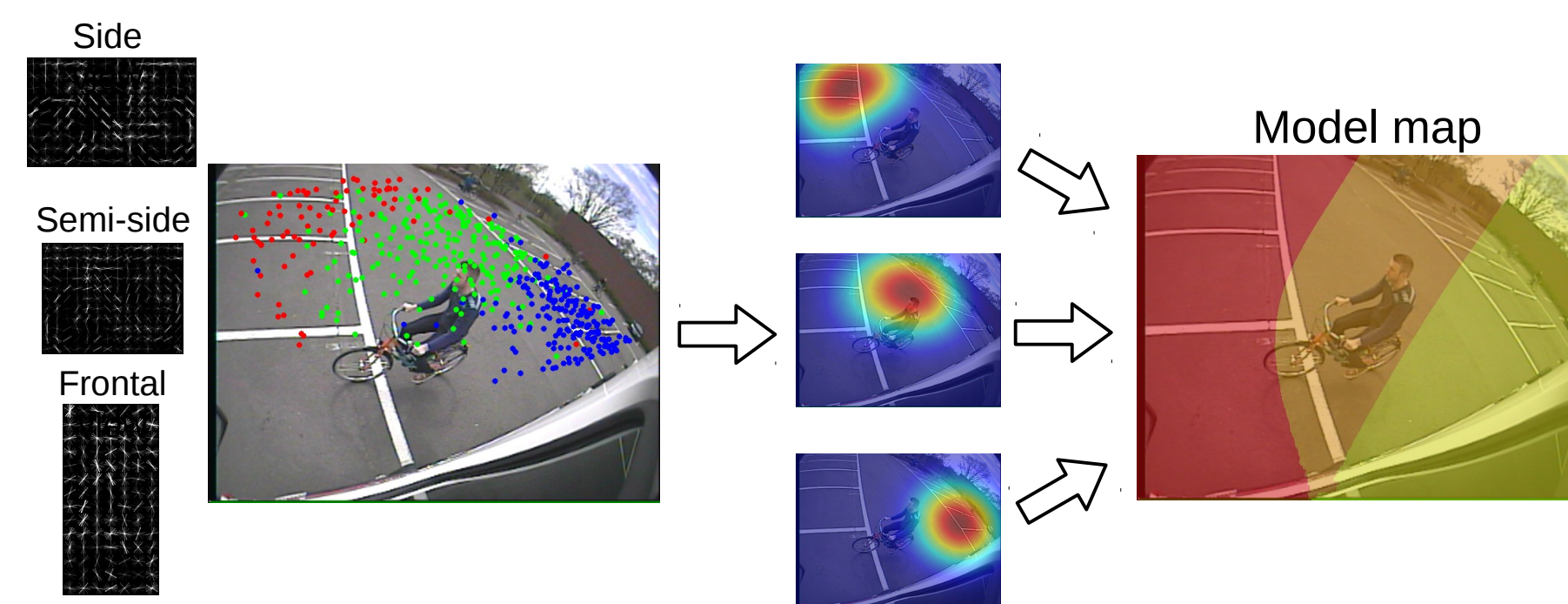
1. Extract and warp patches [15]:

- Assume **flat ground plane**, planar objects
- Modeled as **perspective transformation**
- Known after one-time calibration



2. Object detection pipeline:

- Evaluation on a **single scale only**: motivation to use the **cascaded DPM** detector [10] as baseline
- Extract features for each patch, and run **three detection models**: pedestrian, upper body and bicycle component
- Select **best bicycle component** based on position in the image:



3. Combine probability maps:

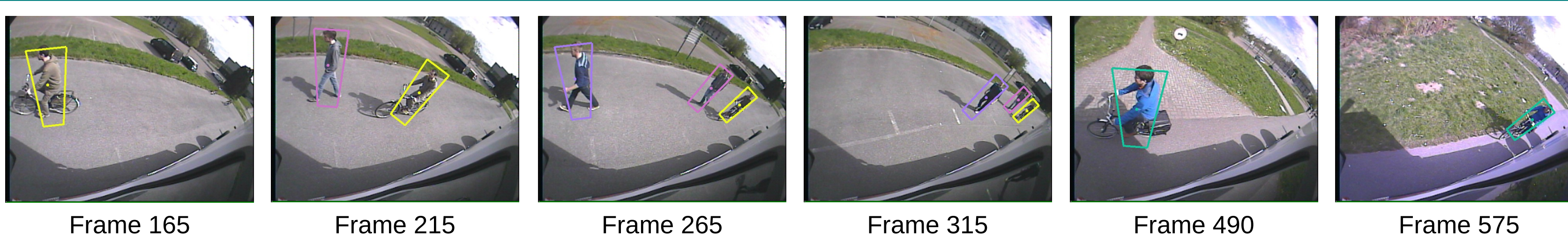
- Shift maps with trained offset and combine probability maps using:

$$P_{final}(x) = \max_{i \in \{1,2,3\}} (P_i(x) - d_i(x)) + G(x) \quad G(x) = \alpha \left[e^{-\frac{x^2}{2\sigma^2}} - 1 \right]$$

4. Map exploration, NMS and tracking:

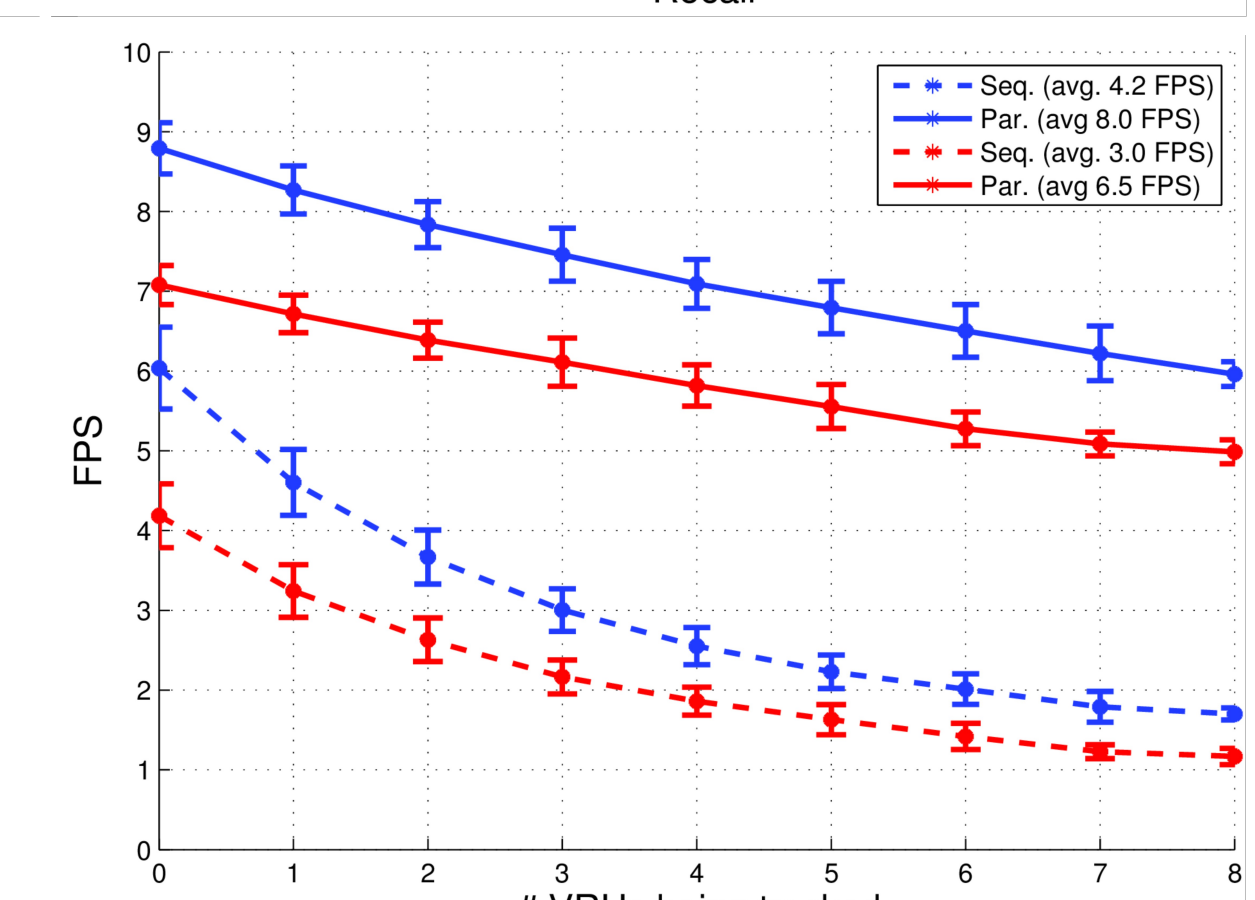
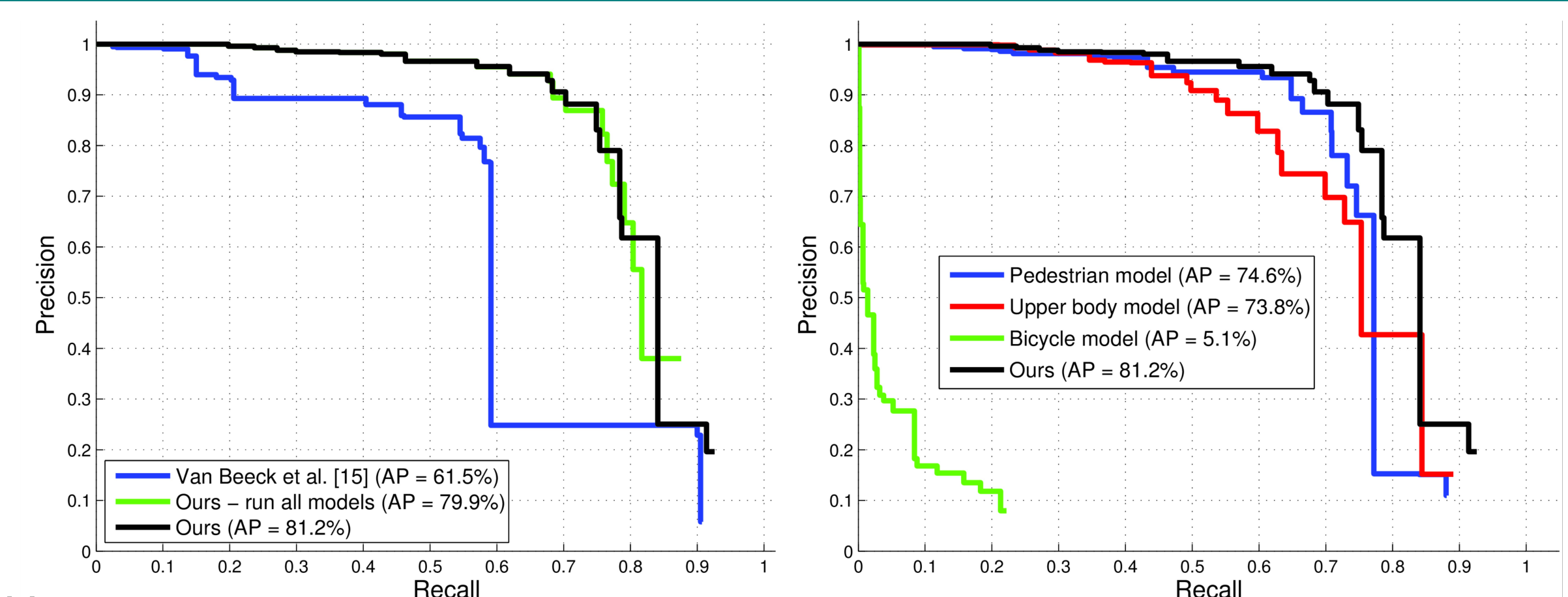
- Integrate in **tracking-by-detection framework**:
 - Default search points** at strategic locations
 - Threshold probability maps** to extract local maxima and perform NMS
 - Instantiate **Kalman filter** for each new detection
 - Constant **velocity model**, predict **future location**
 - Verify** for each detection if **track exists** within radius based on scale, otherwise start **new track**

Experiments & Results



We performed extensive experiments:

- Commercial blind spot camera (Orlaco115) on genuine truck (Volvo FM12)
- Simulated **seven** different **dangerous** blind spot situations
- Total test set about **5000 frames**: 640x480 with 15 fps: 3600 pedestrians and 2400 bicyclists labeled
- Both single core and parallel Implementation, Matlab code with C and OpenCV, Intel Xeon E5 CPU at 3.1 GHz
- We achieve **excellent accuracy** (AP = 81.2%) at an average of **8 fps** (Matlab proof-of-concept implementation)



Conclusion

- We presented a multiclass object tracking framework targeting a specific application: detection and tracking of pedestrians and bicyclists in the blind spot camera of a truck
- Using an **integration** of a **warping window approach** with an **efficient detection scheme** where we only run specific viewpoint detectors
- We achieve **excellent accuracy results**, while keeping the **computational complexity** adequate for practical applications